

# CS 585 Assignment 1

Don Johnson

January 20, 2014

## 1 Assignment

### 1.1 Learning Objectives

### 1.2 Technical Tasks - Relevant Code and Output

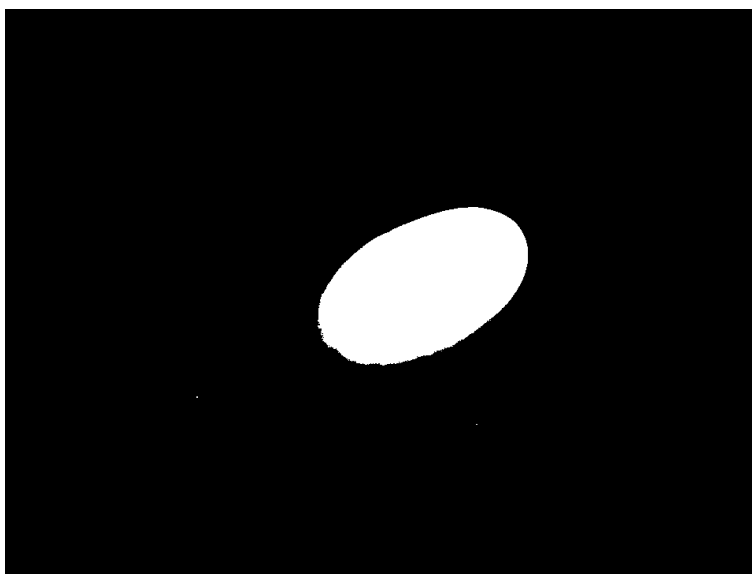


Figure 1: Image Mask for Red Egg at Threshold of 175

Code used to create mask:

```
void thresholdChannel(Mat& image, Mat& mask, int channelIndex, int threshold)
{
    int channels = image.channels();

    for(int row=0; row<image.rows; row++)
    {
        unsigned char* imagePtr = image.ptr<unsigned char>(row);
        unsigned char* channelPtr = mask.ptr<unsigned char>(row);

        for(int col=0; col<image.cols; col++)
        {
            int index = col*channels + channelIndex;
            channelPtr[col] = imagePtr[index] >= threshold ? 255 : 0;
        }
    }
}
```

### 1.3 Questions

	Color/Value	R	G	B
1.	Red	255	0	0
	Yellow	255	255	0
	White	255	255	255

2. Since the red channel value for both a yellow and a red object are the same, the output of the thresholding program, which only looks at the red channel, will be the same.
3. The results were what I expected based on the explanation in the previous item.

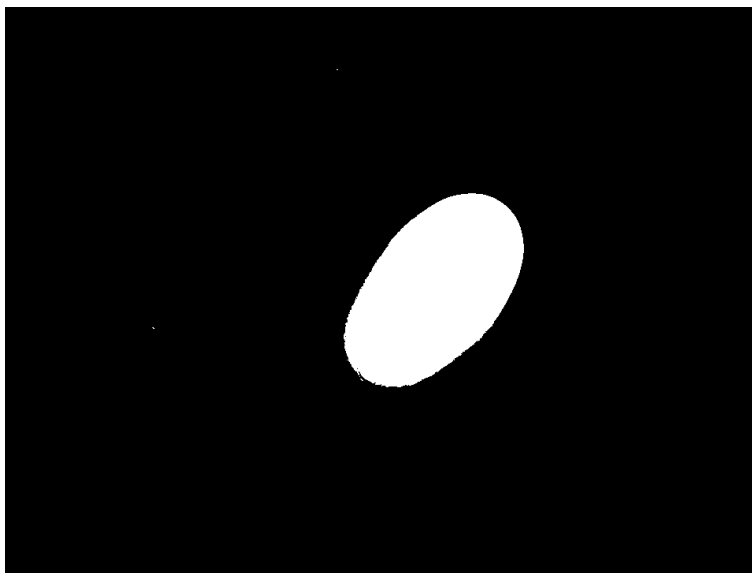


Figure 2: Mask Created by Red Threshold of Yellow Egg

4. For a given pixel with RGB values, and ignoring the blue value, look at the ratio  $R/(R+G)$ . Define "red" to be when the ratio is greater than  $2/3$ . Define "yellow" to be where the ratio is between  $1/3$  and  $2/3$ . Exclude noise and shadow for red by requiring that the sum of red and green be greater than 100 and for yellow that the sum be greater than 300. Using these criteria and assigning 255 to red and 128 to yellow, I created this mask of the red and yellow egg image:

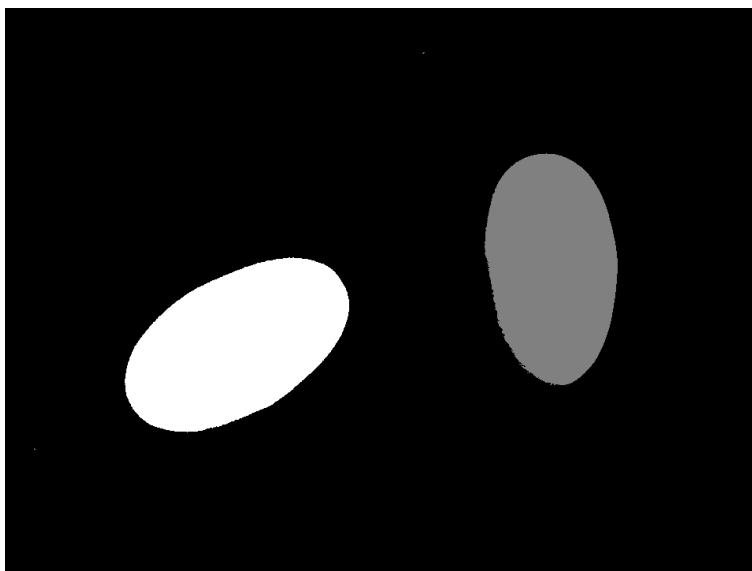


Figure 3: Detecting Red and Yellow Objects Using RGB Ratios (Red=White, Yellow=Gray)

This is a code snippet with the ratio logic:

```

int indexGreen = col*channels + 1; // red
int indexRed = col*channels + 2; // green
int redPlusGreen = imagePtr[indexRed] + imagePtr[indexGreen];
float ratio = imagePtr[indexRed]/(redPlusGreen + 0.1);
channelPtr[col] = ratio > oneThird && redPlusGreen > 300 ? 128 : 0;
channelPtr[col] = ratio > twoThirds && redPlusGreen > 100 ? 255 : channelPtr[col];

```

5. Took a picture of my son's penguin, but against a white background to minimize difficulties. Also decided to try detecting yellow as well as red, but picked up white and gray so had to add two more lines to the previous code to eliminate pixels with blue RGB component greater than 100:

```

int indexBlue = col*channels; // blue
channelPtr[col] = imagePtr[indexBlue] > 100 ? 0 : channelPtr[col];

```

The algorithm labeled the brighter part of the orange feet and my hand as yellow. Here are the original image and output mask:



(a) Original



(b) Mask with Red=White and Yellow=Gray

Figure 4: The Penguin

6. The color representation idea that I found most interesting was the trichromatic theory of perception where monochromatic colors, in terms of perception, can be created as a mixture of three primaries: 700 nm, 546 nm and 436 nm. Also, found the description of the CIE, XYZ color space and the matrix that relates it to the RGB space interesting.
7. A monochromatic image of dimension 960x720 has 691,200 pixels. A RGB image of that size has  $691,000 \times 3 = 2073600$  bytes, but still the same number of pixels.
8. Found these timing values in the original code:

Name	Elapsed Time (sec)
Linear Index v1	67.690270412808275
Linear Index v2	21.070437060547910
Linear Index v3	3.7960514995024437
Row Ptr	3.5284727421810489

In the version 3, addOne.LinearIndex() function, the image.step[0] function is only being called 720 times, once per row. In the version 2, the image.step[0] function is being called 2,073,600 times, once per each channel in every pixel. Since image.step[0] always returns 2880, the number of bytes in a row, it can be moved outside of the loops in both versions of the addOne.LinearIndex() function. When you do this, the run times are almost identical:

Name	Elapsed Time (sec)
Linear Index v2	3.9606397209834103
Linear Index v3	3.8033933929887844

Conclusion, run times differences between v2 and v3 of the addOne.LinearIndex() function are caused by calls to the image.step[0] function and not by differing number of mathematical operations. As a side note, and not counting the pixel increment, v2 has two multiplications and two additions while v3 has only one multiplication and two additions per pixel. In the rainbowSpheres.jpg image we are using, there a 691,200 pixels so the various operation just mentioned happen one or two times that number. The slightly slower time of v2, once the step[] function has been removed may be caused by the extra multiplication it has over v3.

- This code snippet below demonstrates the strange thing that is happening in when we add 1 to a color channel of an image. If the color channel is saturated (equal to 255) for a particular pixel, the color channel value rolls over to 0 because it is a unsigned byte. This why the more saturated areas of the image showed a strange speckled effect. The color also changes. For example if you add one to a saturated green channel in a yellow pixel, the pixel changes to red.

Here is the code and output:

```
unsigned char nvar = 253;
for (int i=0; i<5; i++){
    printf("nvar + 1: %d\n", (unsigned int)++nvar);
}
```

Output:

```
nvar + 1: 254
nvar + 1: 255
nvar + 1: 0 (Here is where the value rolled over)
nvar + 1: 1
nvar + 1: 2
```

## 2 Lecture preparation

### 2.1 Preparation for Lecture 2

- Formula for the area A, of the base of a cone, given an apex angle  $\alpha$ , and height h:

$$\begin{aligned}\tan \alpha &= \frac{r}{h} \\ r &= h \tan \alpha \\ A &= \pi r^2 \\ A &= \pi (h \tan \alpha)^2\end{aligned}$$

- Solving for side Z of the blue triangle using properties of similar triangles:

$$\begin{aligned}\frac{Z}{f} &= \frac{X}{u} \\ Z &= \frac{X}{u} \cdot f \\ Z &= \frac{5000mm \cdot 25mm}{0.540mm} \\ Z &= 231.48m\end{aligned}$$

3. Solving for side u of the red triangle using properties of similar triangles:

$$\begin{aligned}\frac{Z}{f} &= \frac{X}{u} \\ u &= \frac{f}{X} \cdot Z \\ u &= \frac{25mm \cdot 1000mm}{10000mm} \\ u &= 2.5mm\end{aligned}$$

4. Count N in pixels of u in for an  $18\mu m$  pixel is:

$$\begin{aligned}N &= \frac{u}{\text{pixels size in } \mu m} \\ N &= \frac{2500\mu m}{18\mu m} \\ N &= 139\text{pixels}\end{aligned}$$

5. The magnitude of the vectors, n, v:

$$\begin{aligned}|n| &= \sqrt{(1)^2 + (1)^2} = \sqrt{2} \\ |v| &= \sqrt{(1)^2 + (0.5)^2} = \frac{\sqrt{5}}{2}\end{aligned}$$

The corresponding unit vectors  $\|n\|, \|v\|$ :

$$\begin{aligned}\|n\| &= \left(\frac{1}{\sqrt{2}}, \frac{1}{\sqrt{2}}\right) \\ \|v\| &= \left(\frac{2}{\sqrt{5}}, \frac{1}{\sqrt{5}}\right)\end{aligned}$$

6. The dot product of unit vectors  $\|n\|, \|v\|$ :

$$\begin{aligned}\|n\| \cdot \|v\| &= \frac{1}{\sqrt{2}} \cdot \frac{2}{\sqrt{5}} + \frac{1}{\sqrt{2}} \cdot \frac{1}{\sqrt{5}} \\ \|n\| \cdot \|v\| &= \frac{3}{\sqrt{10}}\end{aligned}$$

7. The angle  $\theta$  between vectors n, v:

$$\begin{aligned}\cos \theta &= \frac{n \cdot v}{|n| \cdot |v|} \\ \cos \theta &= \frac{1 \cdot 1 + 1 \cdot \frac{1}{2}}{\sqrt{2} \cdot \frac{\sqrt{5}}{2}} \\ \cos \theta &= \frac{3}{\sqrt{10}} \\ \theta &= 18.44^\circ\end{aligned}$$

8. The projection p of v, onto n:

$$\begin{aligned}p &= |v| \cos \theta \cdot \|n\| \\ p &= \frac{\sqrt{5}}{2} \cdot \frac{3}{\sqrt{10}} \left(\frac{1}{\sqrt{2}}, \frac{1}{\sqrt{2}}\right) \\ p &= \frac{3}{2\sqrt{2}} \left(\frac{1}{\sqrt{2}}, \frac{1}{\sqrt{2}}\right) \\ p &= \left(\frac{3}{4}, \frac{3}{4}\right)\end{aligned}$$

9. The perpendicular vector  $q$ , between  $v$  and  $n$ :

$$\begin{aligned} q &= v - p \\ q &= (1, \frac{1}{2}) - (\frac{3}{4}, \frac{3}{4}) \\ q &= (\frac{1}{4}, -\frac{1}{4}) \end{aligned}$$

Finally,  $r$  equals the sum of the components we just calculated:

$$\begin{aligned} r &= p + q \\ r &= (\frac{3}{4}, \frac{3}{4}) + (\frac{1}{4}, -\frac{1}{4}) \\ r &= (\frac{1}{2}, \frac{1}{2}) \end{aligned}$$

## 2.2 Preparation for Lecture 3

1. Adapted from source: [http://www.csl.mtu.edu/cs2321/www/newLectures/26\\_Depth\\_First\\_Search.html](http://www.csl.mtu.edu/cs2321/www/newLectures/26_Depth_First_Search.html)

Definitions:

$G$  is a graph

$G.\text{incidentEdges}(v)$  – all of the edges associated with vertex  $v$

$G.\text{opposite}(v, e)$  – a vertex at the other end of edge  $e$  connected to  $v$

Note: Every edge and vertex is marked initially as unexplored.

The algorithm terminates when all vertices and edges have been explored.

The initial vertex is the root.

Algorithm **for** depth-first, recursive transversal of graph  $G$ :

DFS(graph  $G$ , Vertex  $v$ )

**for** all edges  $e$  in  $G.\text{incidentEdges}(v)$  **do**

**if** edge  $e$  is unexplored then

$w = G.\text{opposite}(v, e)$

**if** vertex  $w$  is unexplored then

            label  $e$  as discovery edge

            recursively call DFS( $G, w$ )

**else**

            label  $e$  as a back edge

2. Algorithm to find connected components of graph and using depth-first transversal algorithm in previous question:

Given a set of vertices  $\{a, b, c, d, e, \dots\}$

Create a 2-dimensional matrix where the row and columns are the labeled with the vertices. A cell value of 1 means they are connected and 0 unconnected.

Initialize  $(a, a), (b, b), (c, c), \dots$  equal to 1 – every vertex is connected to itself.

**for** all vertices  $v$  in  $G$

    Do DFS( $G, v$ ) only **for** sum of row and column with same label is greater than one

**for** each  $v$ , treated as root vertex, set connection matrix cell  $(v, v_1), (v, v_2)$



Figure 5: Results from running my program from section 1.2 on the picture "twoRedEggs.jpg"

- 3.
4.
  - To find the number of red objects, I would define red for a pixel as a minimum ratio between the red channel and the blue, green channels
  - Create a mask of all of those pixels defined as red, setting them equal to 1 and all other pixels equal to 0
  - Run a depth first algorithm against the mask, visiting only the pixels equal to 1, treating each pixel as a vertex and each pixel boundary as an edge. Make it a greedy algorithm, collecting sets of connected pixels and keeping track of which pixels have been visited.
  - Count the pixel sets to determine number of red objects
5. My strategy for discarding the small extra regions that may pass through your red-object detector would be to require a minimal set size when counting the sets in the last step of the algorithm outlined in the previous question.
6. If I was trying to create an rordered list of pixels that comprised the perimeter of an image region, for example, one of the red objects in the previous problem, I would:
  - Assume a x,y coordinate system for the image with (0,0) being the upper-left hand coordinate and both x and y increasing as you went down and to the right of the image.
  - Find a pixel in one of the objects pixels sets from the previous problem with a value equal to 1 and having an x-coordinate minima in the set of pixels for that object. This will be one of the far left pixels in the object.
  - Check for neighbor pixels with a value of one in this order: left, left-up, up, right-up, right, right-down, down, left-down
  - When a non-zero neighbor is found, add it to a linked list, mark the pixel as visited and repeat the left to clockwise search pattern starting at the pixel boundary at which you entered.
  - Repeat the process until you return to the first pixel where you started. The approach is similar to the approach used by bright physician from Egypt on the Maze of the Minotaur. The linked list contains the boundary pixels for the object.